

Machine Learning and Artificial Intelligence

“Programming agents by hand can be very tedious; some more expeditious method seems desirable.”

—Alan Turing, 1950

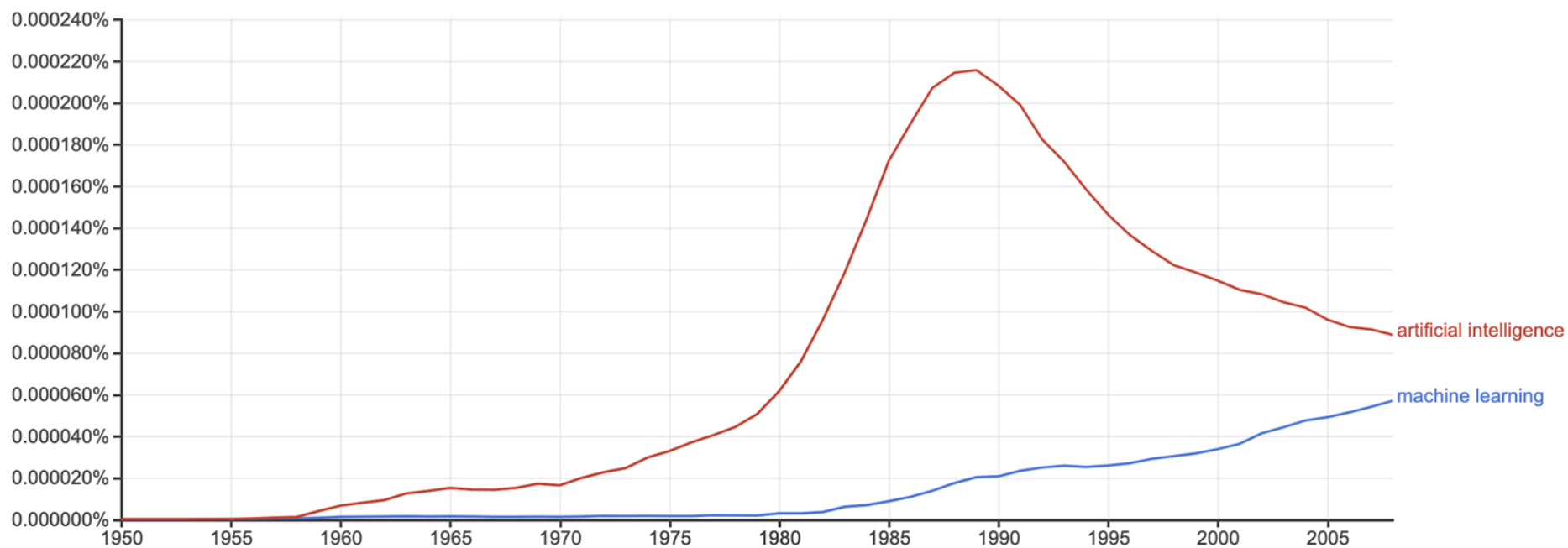
Google Books Ngram Viewer

Graph these comma-separated phrases: machine learning,artificial intelligence

☐ case-insensitive

between 1950 and 2008 from the corpus English (2012) with smoothing of 3

[Search lots of books](#)





● machine learning
Search term

● artificial intelligence
Search term

+ Add comparison

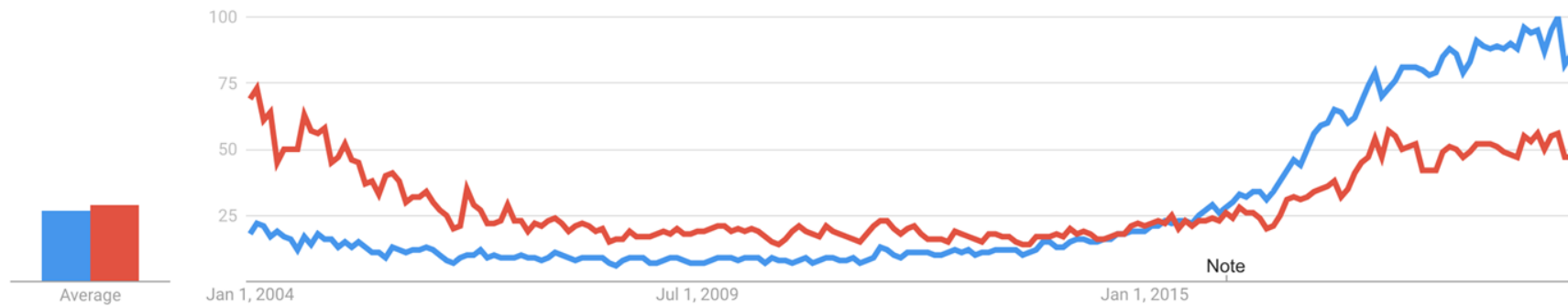
Worldwide ▼

2004 - present ▼

All categories ▼

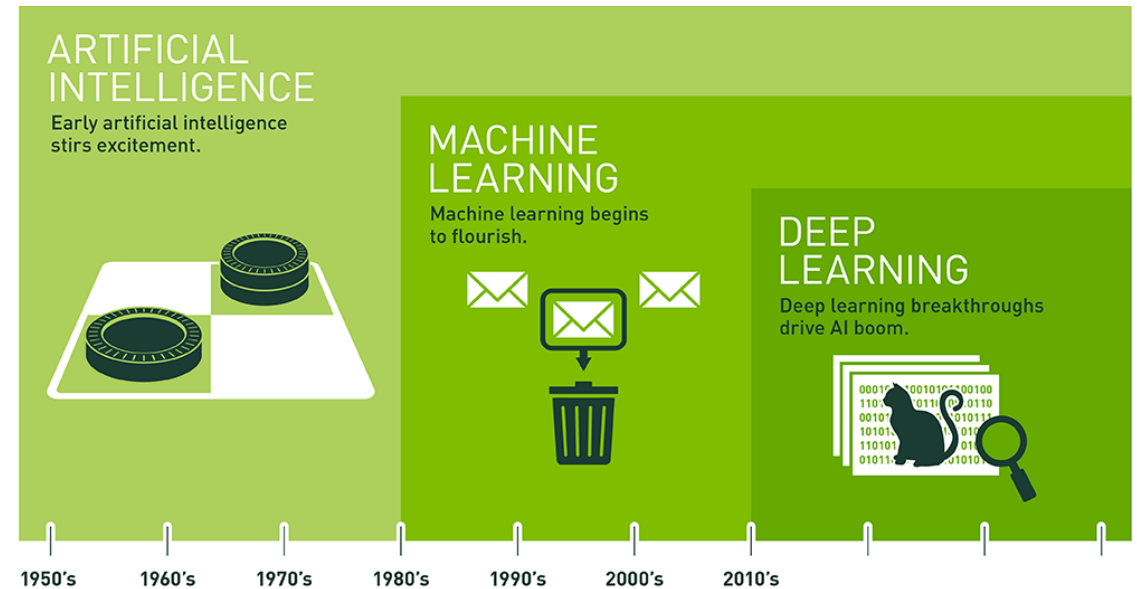
Web Search ▼

Interest over time ?



ML and AI

- Artificial Intelligence
 - Computational rationality
 - Decision making under uncertainty
- Machine Learning (ML)
 - Finding patterns in data



Since an early flush of optimism in the 1950s, smaller subsets of artificial intelligence – first machine learning, then deep learning, a subset of machine learning – have created ever larger disruptions.

Autonomous Driving



Autonomous Driving

Object identification

Object detection

Object tracking/prediction

Route planning

Fully autonomous driving

Autonomous Driving

Artificial Intelligence

Object identification

Object detection

Object tracking/prediction

Route planning

Fully autonomous driving

Autonomous Driving

Object identification	Image → Label
Object detection	Image → Object bounding boxes
Object tracking/prediction	Video → Moving bounding boxes
Route planning	Video, Map, Coordinates → Actions
Fully autonomous driving	Video, Map, Coordinates, Ethics, ... → Actions

Autonomous Driving

Classical Machine Learning

Object identification	Image → Label
Object detection	Image → Object bounding boxes
Object tracking/prediction	Video → Moving bounding boxes
Route planning	Video, Map, Coordinates → Actions
Fully autonomous driving	Video, Map, Coordinates, Ethics, ... → Actions

Supervision

- Supervised learning (inputs, labels)
- Unsupervised learning (inputs)
- Reinforcement learning (inputs, eventual rewards)

Machine Learning and Artificial Intelligence

The End

Course Overview

Machine Learning Is Going Mainstream

- “Data is the new oil.”
- “AI is the new electricity.”
- “Privacy is the new luxury.”
- “Machine learning is the future.”

Course Goals

- Practical introduction to the machine learning field
- Building blocks and theory of neural networks
- Useful programming tools (including Tensorflow)
- Learn how to stay current as the field changes

Python Notebooks

- Weekly notebooks are the interactive textbook of this course
- Options for running the code
 - Run locally using Jupyter (need to install libraries)
 - Run using Google's Colab (everything is installed)

Course Work

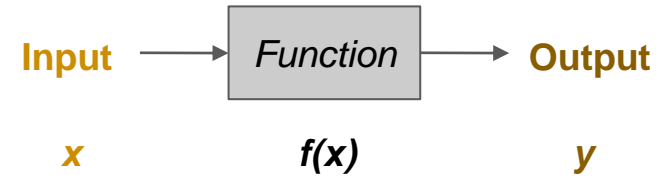
- Weekly notebook exercises
- Final project
 - Small groups
 - Choose from approximately three Kaggle competitions
 - Run experiments
 - Write a summary report and present to class

Course Overview

The End

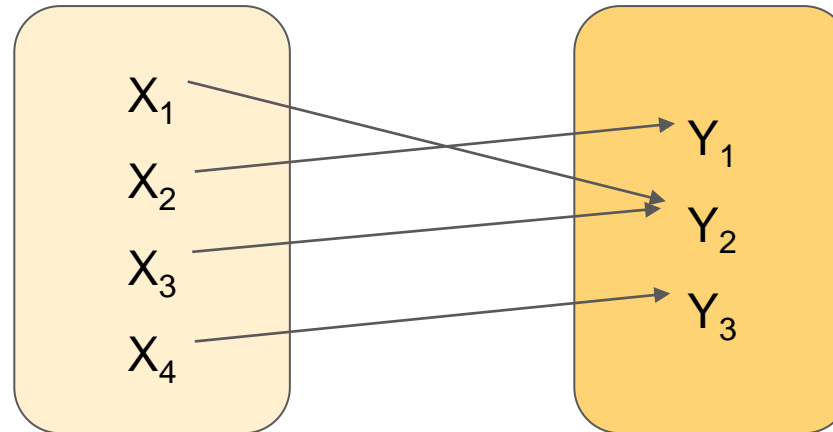
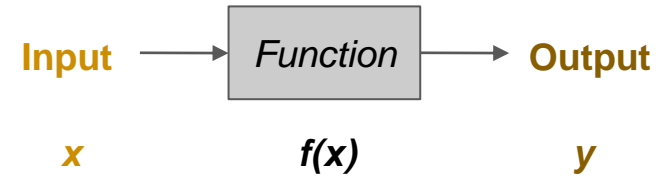
What Is a Function?

What Is a Function?



What Is a Function?

Mapping between sets: for each input item, there is one corresponding output item.



Function Examples

Function Examples

- `add_six(number)`
- `get_prime_factors(number)`
- `upper_case(string)`
- `convert_time(loc1, time, loc2)`
- `is_it_raining(weather_report)`

Function Examples

- `add_six(number)`
- `get_prime_factors(number)`
- `upper_case(string)`
- `convert_time(loc1, time, loc2)`
- `is_it_raining(weather_report)`
- `height_from_shoe_size(shoe_size)`
- `is_cat(image)`
- `is_positive_movie_review(text)`
- `text_from_audio(audio)`
- `will_it_rain_tomorrow(weather_report)`

Learned Functions (Models)

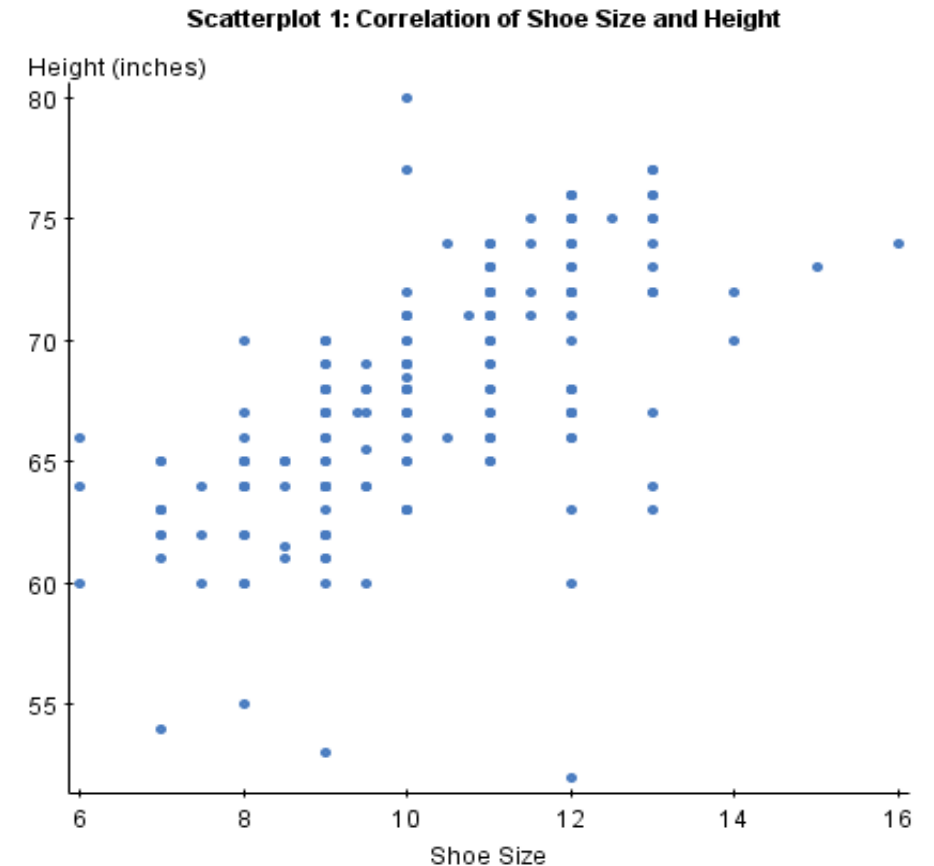
How would you write this function:

`height_from_shoe_size(shoe_size)`

Learned Functions (Models)

How would you write this function:
`height_from_shoe_size(shoe_size)`

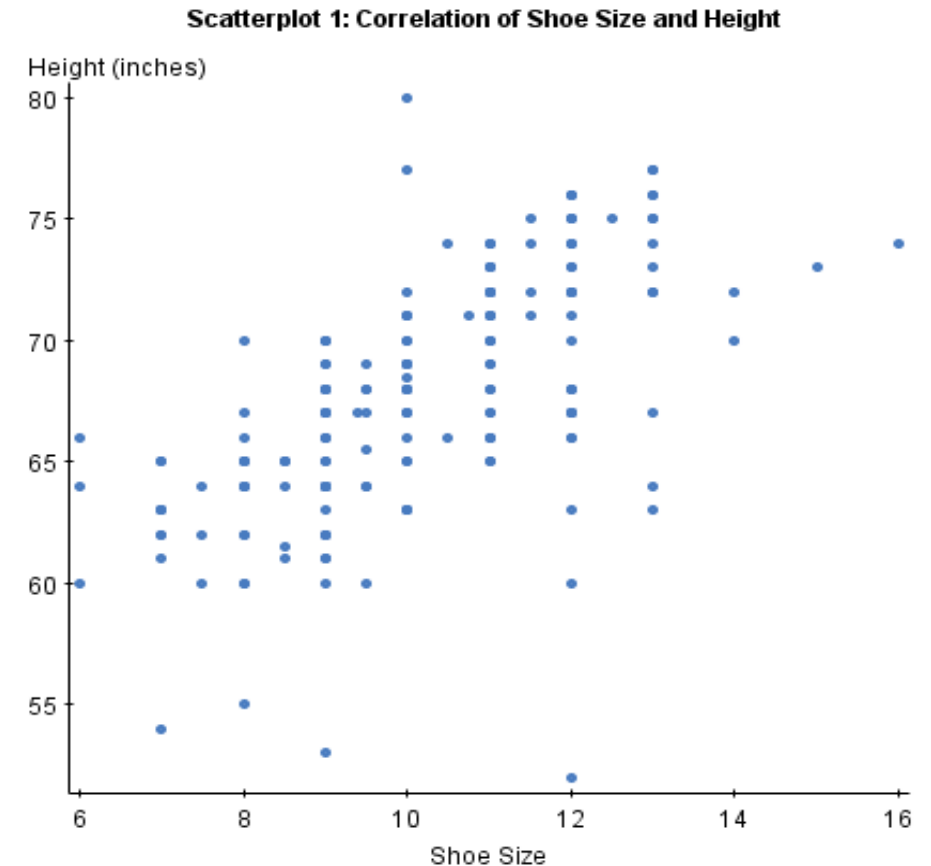
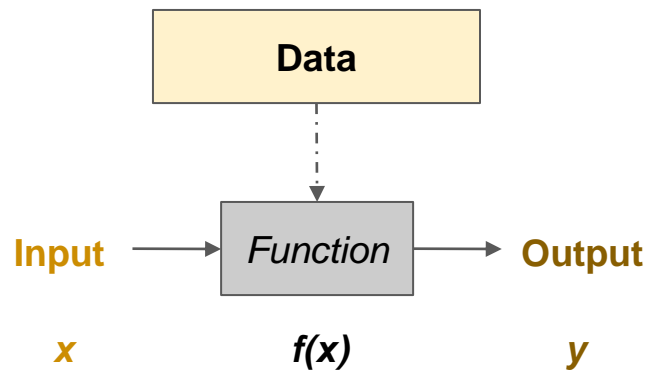
Does this help?



Learned Functions (Models)

How would you write this function:
`height_from_shoe_size(shoe_size)`

Does this help?



What Is a Function?

The End

Function Testing

Function Testing

```
def add_six_test():  
    ???
```

Function Testing

```
def add_six_test():  
    assertEquals(add_six(-3),  
3)  
    assertEquals(add_six(3), 9)
```

- Testing logic
- Check edge cases

Function Testing

```
def add_six_test():  
    assertEquals(add_six(-3),  
3)  
    assertEquals(add_six(3), 9)
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```
def height_from_shoe_size_test():  
    ???
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- Testing logic
- Check edge cases

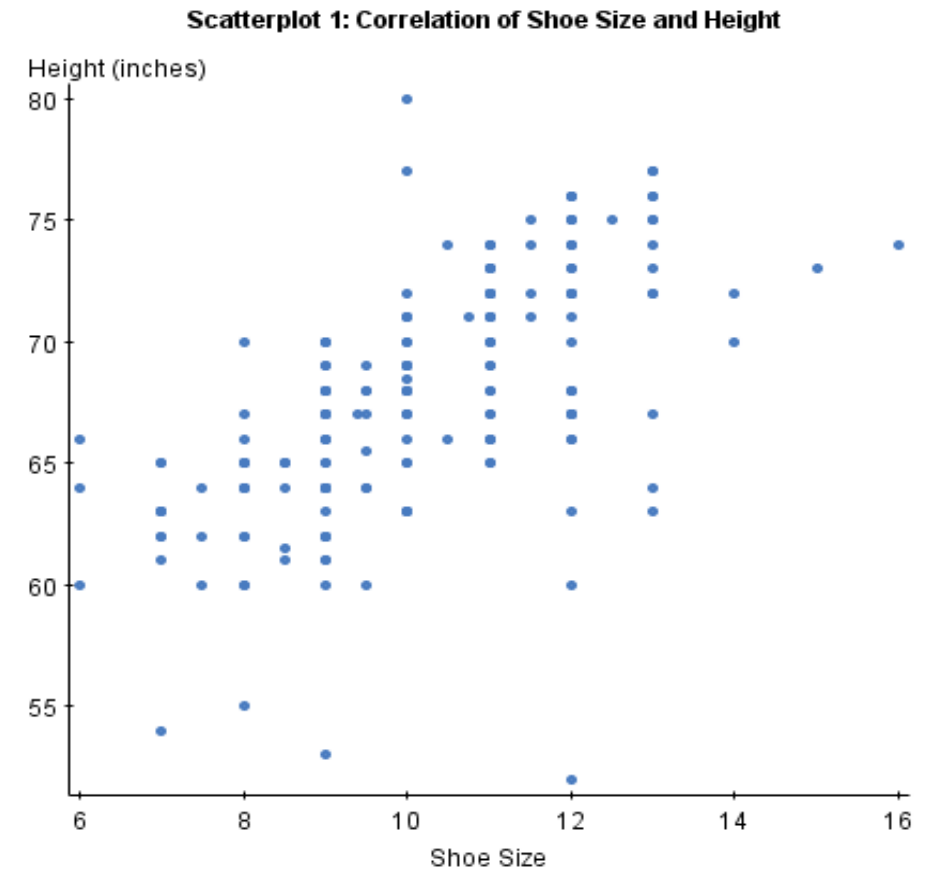
Function Testing

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def height_from_shoe_size_test():  
    ???
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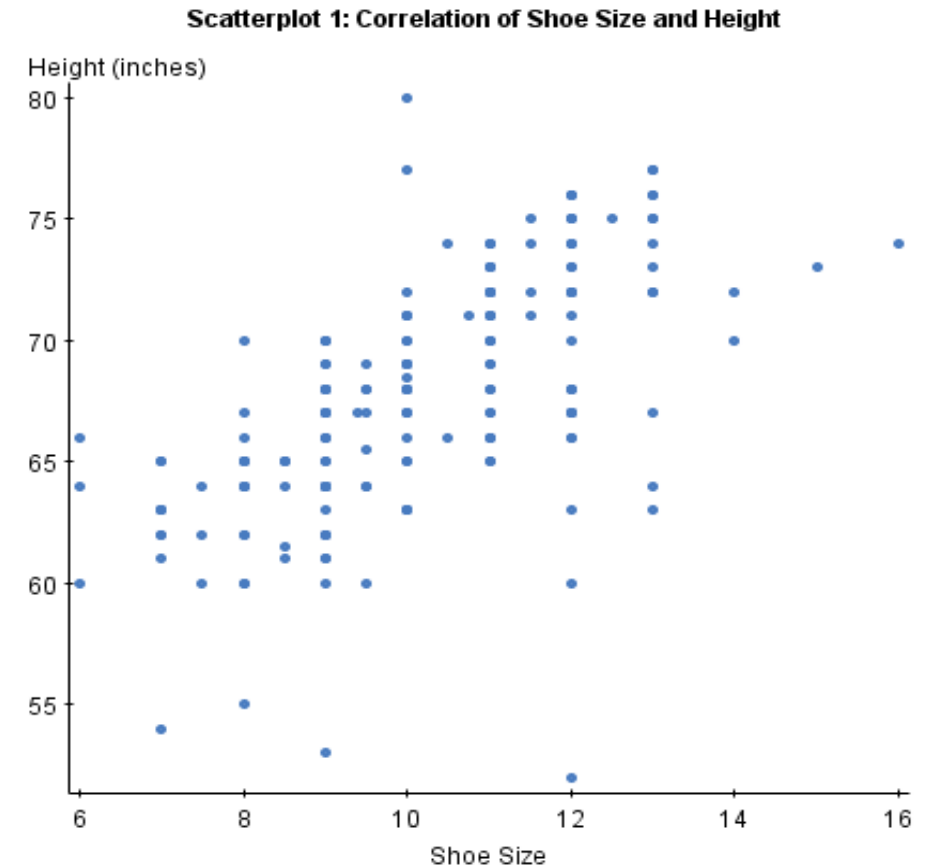
- Testing logic
- Check edge cases
- Statistical testing
- Check average performance

Evaluating Performance



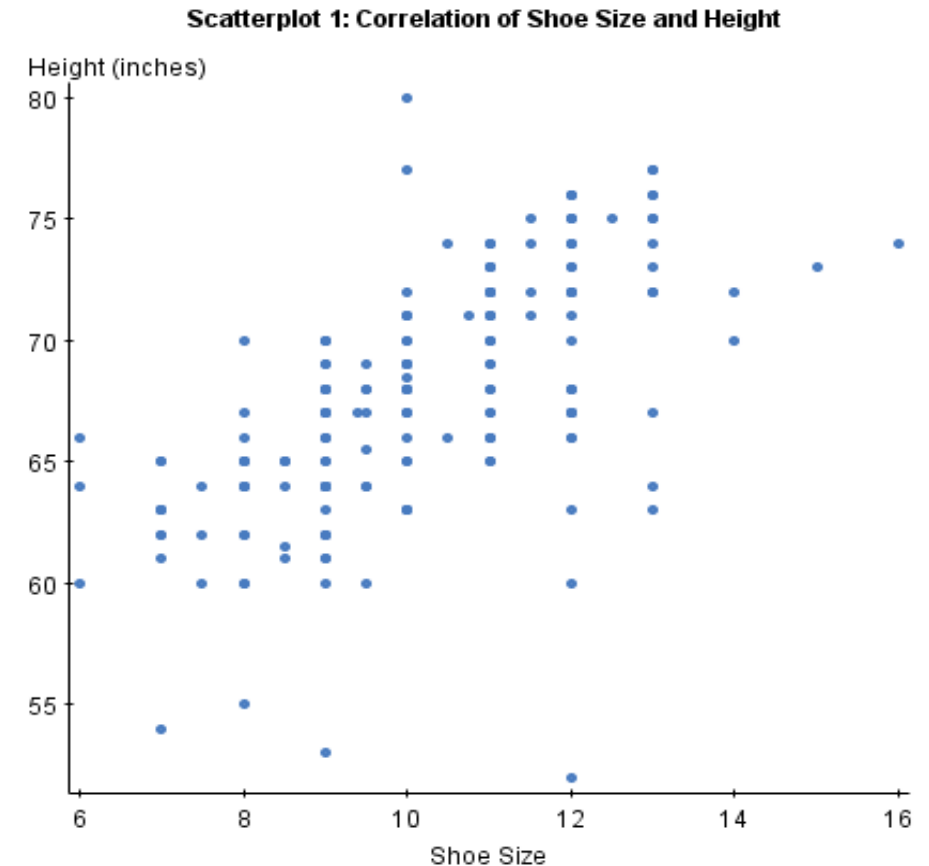
Evaluating Performance

- Need labeled examples
 - Size: 9, Height: 61
 - Size: 12, Height: 66
 - Etc.



Evaluating Performance

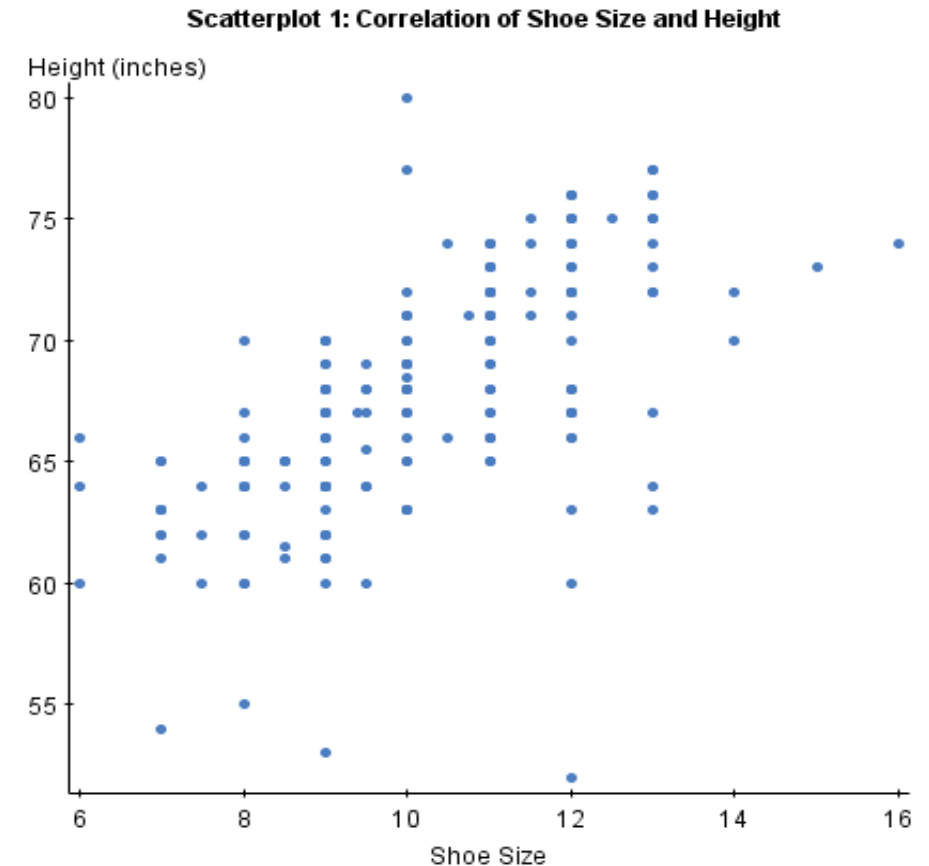
- Need labeled examples
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 - Etc.
- Compare function output (predictions) to labels
 - “Error” or “Loss” or “Cost”



Evaluating Performance

- Need labeled examples
 - Size: 9, Height: 61
 - Size: 12, Height: 66
 - Etc.
- Compare function output (predictions) to labels
 - “Error” or “Loss” or “Cost”

$$MSE = \frac{1}{|Y|} \sum_{y_i \in Y} (y_i - \hat{y}_i)^2$$



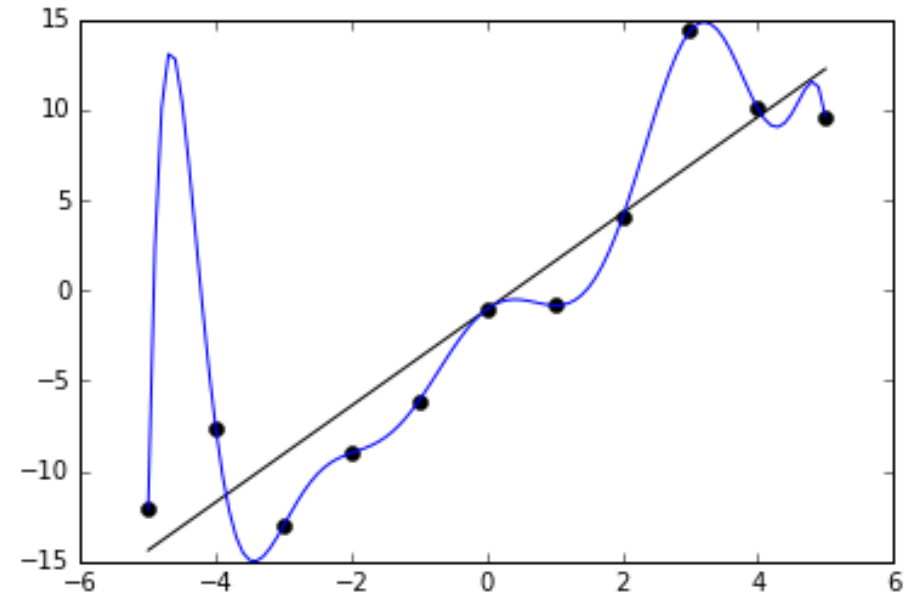
Function Testing

The End

Generalization

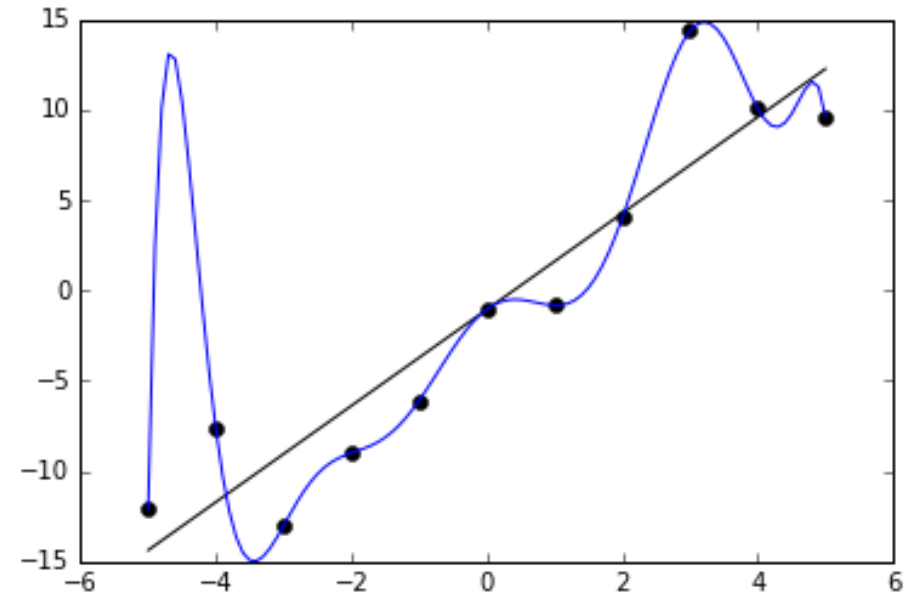
Predictions

- Suppose the points are the labeled examples



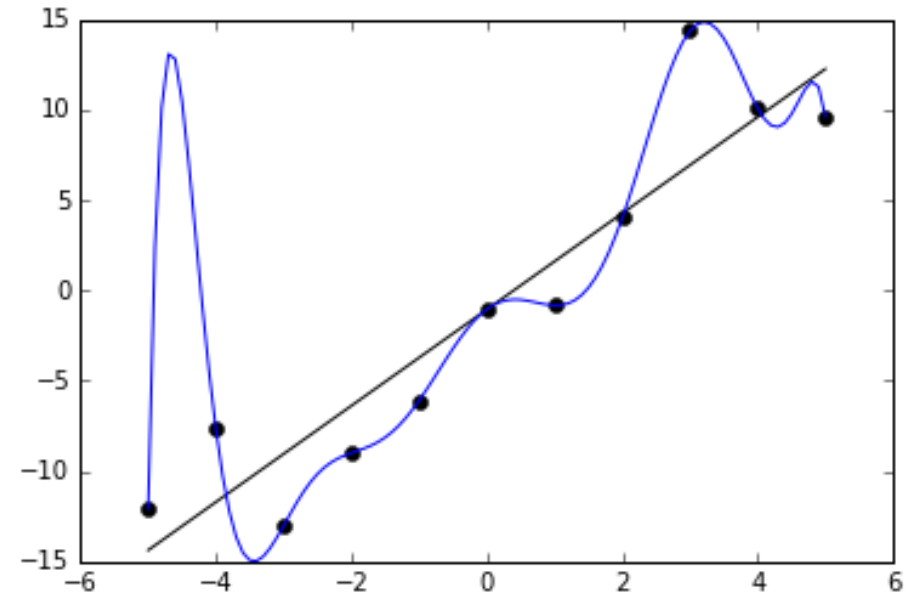
Predictions

- Suppose the points are the labeled examples
 - Are both black and blue lines possible models?



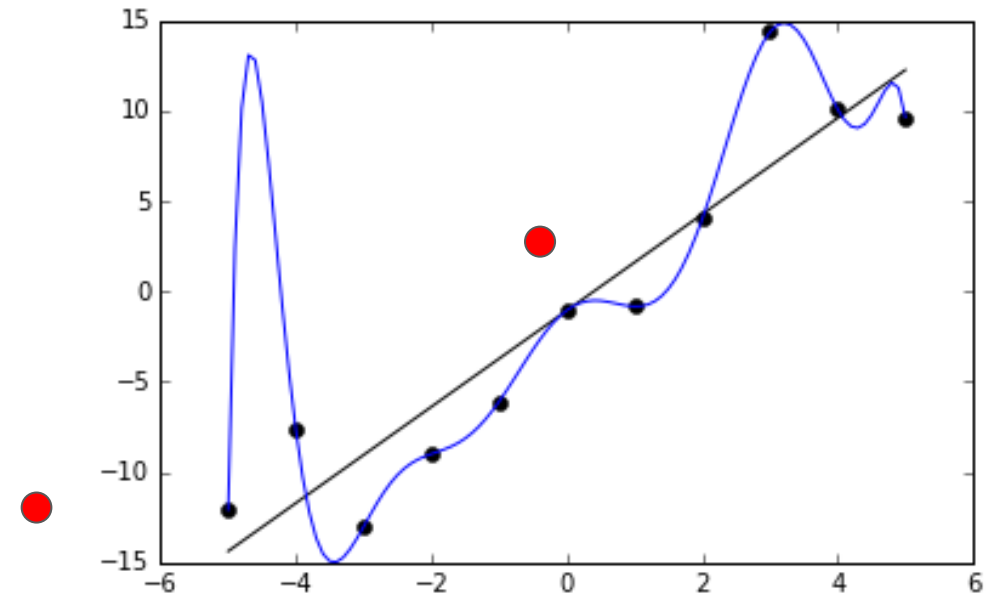
Predictions

- Suppose the points are the labeled examples
 - Are both black and blue lines possible models?
 - Which is better?



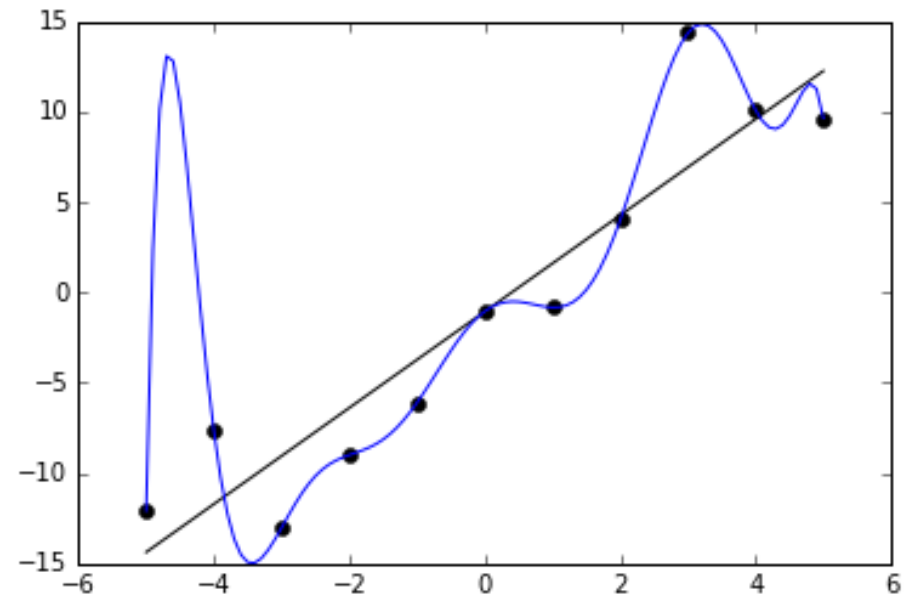
Predictions

- Suppose the points are the labeled examples
 - Are both black and blue lines possible models?
 - Which is better?
 - How do they perform with some new data?



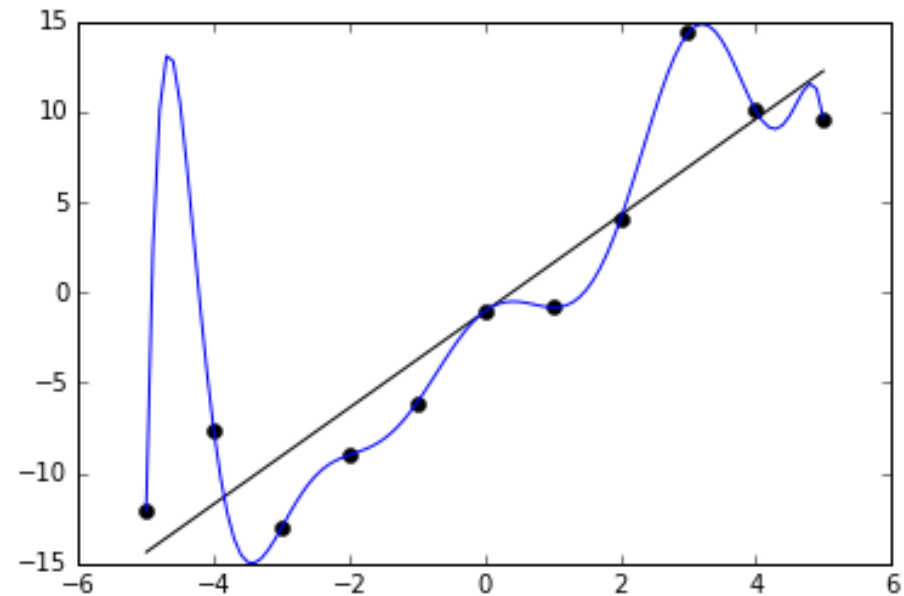
The Train/Test Split

- Models are useful for making new predictions
 - This is called ***generalization***



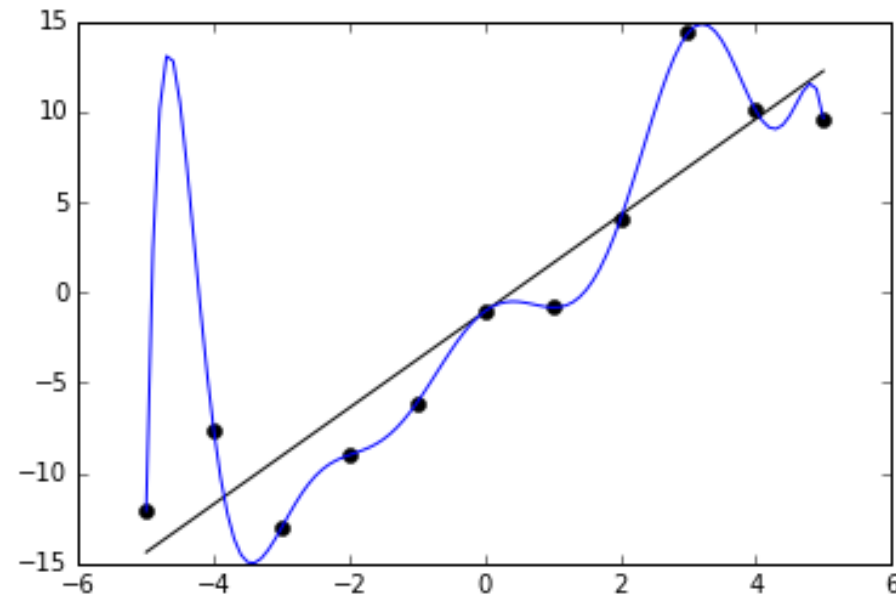
The Train/Test Split

- Models are useful for making new predictions
 - This is called ***generalization***
- Simulate this by splitting data into train and test
 - Could be a random split
 - Could depend on other properties of the data



The Train/Test Split

- Models are useful for making new predictions
 - This is called **generalization**
- Simulate this by splitting data into train and test
 - Could be a random split
 - Could depend on other properties of the data
- Blue model is an example of **overfitting**



Generalization

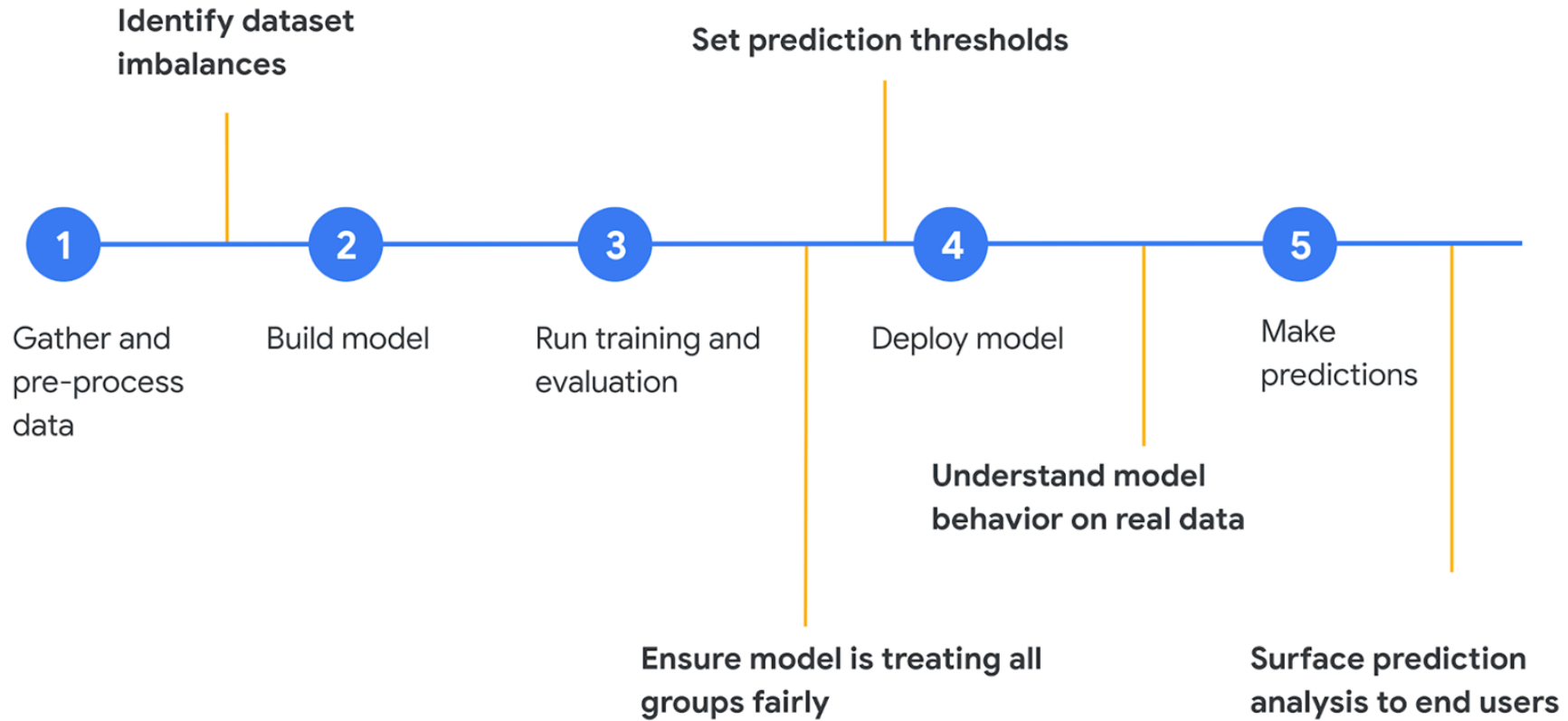
The End

Machine Learning Framing

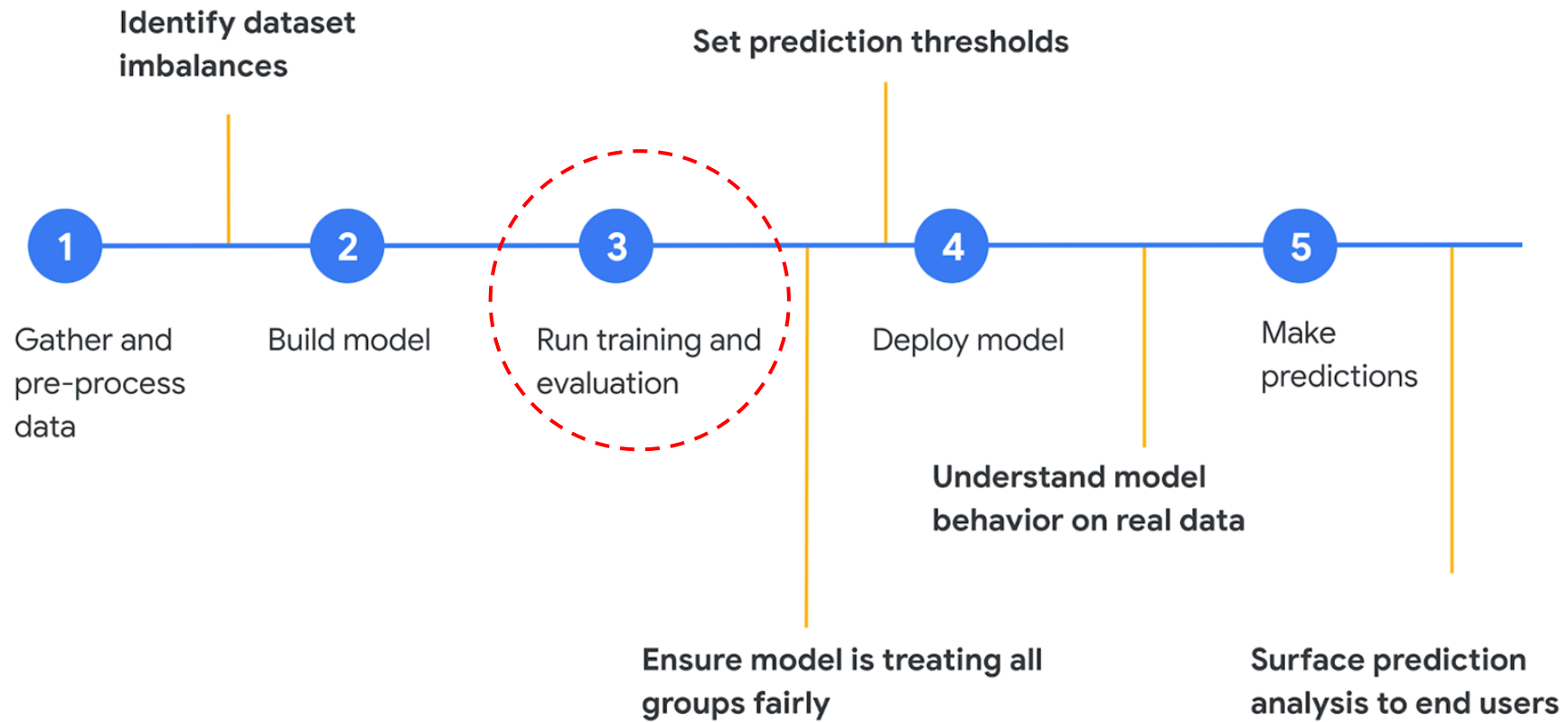
Summary So Far

- You're already familiar with *logical* functions
 - Outputs are typically deterministic
 - Testing is logical and typically checks extreme cases
- ML is about learning *statistical* functions from data
 - Outputs are predictions
 - Testing is statistical and typically checks average case
- ML depends on labeled data
 - Behavior of the model reflects the data
 - Need separate test data to evaluate generalization

Machine Learning Framing



Machine Learning Framing



Framing: Defining a Task

- What are the inputs and outputs?
- What is the labeled data?
- Considerations for train/test split?
- How will your function be used?
- How will you evaluate predictions?
- What is a baseline predictor?

Machine Learning Framing

The End

Lung Cancer Screening

Example One

Lung Cancer Screening

- What are the inputs and outputs?
- What is the labeled data?
- Considerations for train/test split?
- How will your function be used?
- How will you evaluate predictions?
- What is a baseline predictor?



Google

What are the inputs and outputs?

- Input: CT images
- Output: cancer probability
- Output: image regions and probabilities



Google

What is the labeled data?

- Scans annotated by human experts



Google

Considerations for train/test split?

- No patient with scans in both train and test



Google

How will your function be used?

- By doctors
- By patients
- By insurance companies
- Does training population match usage population?



Google

How will you evaluate predictions?

- Classification accuracy



Google

What is a baseline predictor?

- Always predict 'no cancer'



Google

Lung Cancer Screening







The End

Sports Outcomes

Example Two







Basketball Predictions

- What are the inputs and outputs?
- What is the labeled data?
- Considerations for train/test split?
- How will your function be used?
- How will you evaluate predictions?
- What is a baseline predictor?

7 p.m. Eastern		RAPTOR SPREAD	WIN PROB.	SCORE
	Pelicans	- 2 . 5	60%	
	Pistons		41%	
8 p.m.				
	Thunder		43%	
	Timberwolves	- 2	57%	
10:30 p.m.				
	Cavaliers		10%	
	Lakers	- 13 . 5	90%	







What are the inputs and outputs?

- Inputs: team1 and team2
- Outputs: win/loss
- Outputs: point difference

7 p.m. Eastern		RAPTOR SPREAD	WIN PROB.	SCORE
	Pelicans	-2.5	60%	
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10:30 p.m.				
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What is the labeled data?

- Previous games and results (from the current season)

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Considerations for train/test split?



- Use past games to predict future results (rolling predictions)



7 p.m. Eastern		RAPTOR SPREAD	WIN PROB.	SCORE
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How will your function be used?

- Gambling: setting lines







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How will you evaluate predictions?

- Accuracy, MSE

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What is a baseline predictor?

- Team win/loss records

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Sports Outcomes

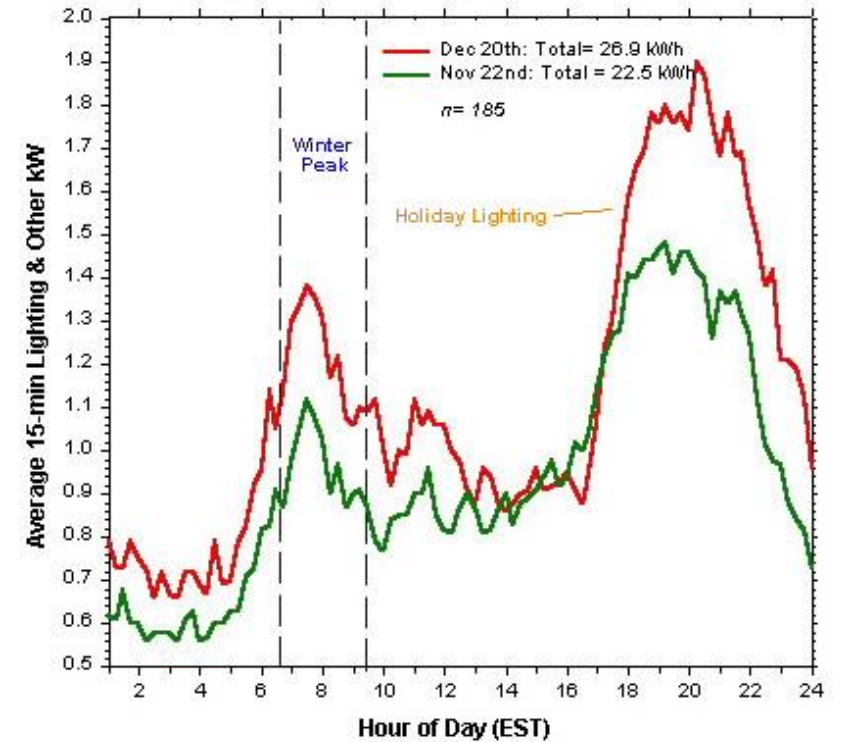
The End

Energy Usage

Example Three

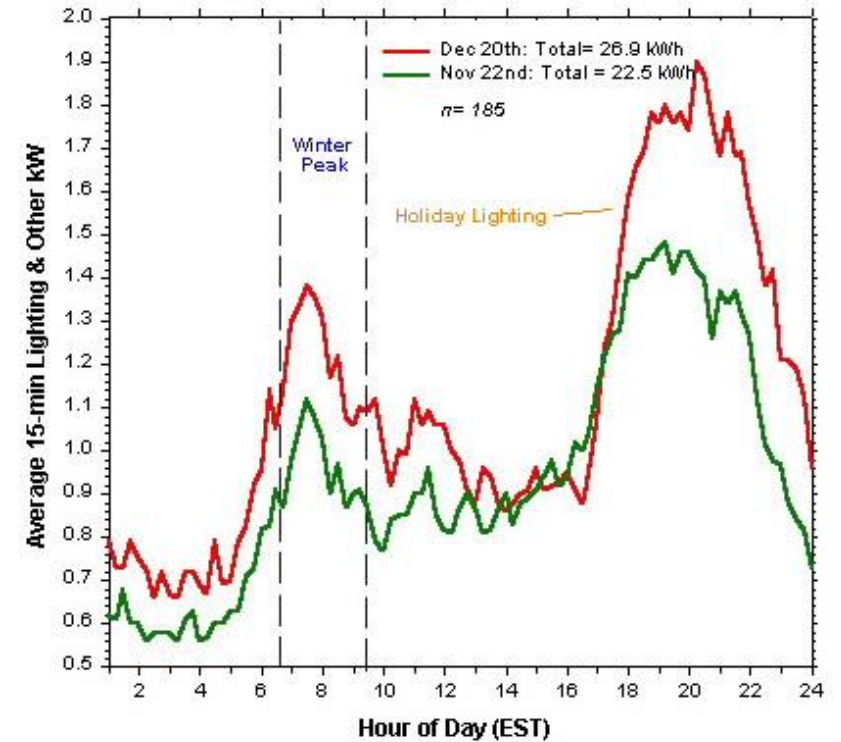
Energy Usage

- What are the inputs and outputs?
- What is the labeled data?
- Considerations for train/test split?
- How will your function be used?
- How will you evaluate predictions?
- What is a baseline predictor?



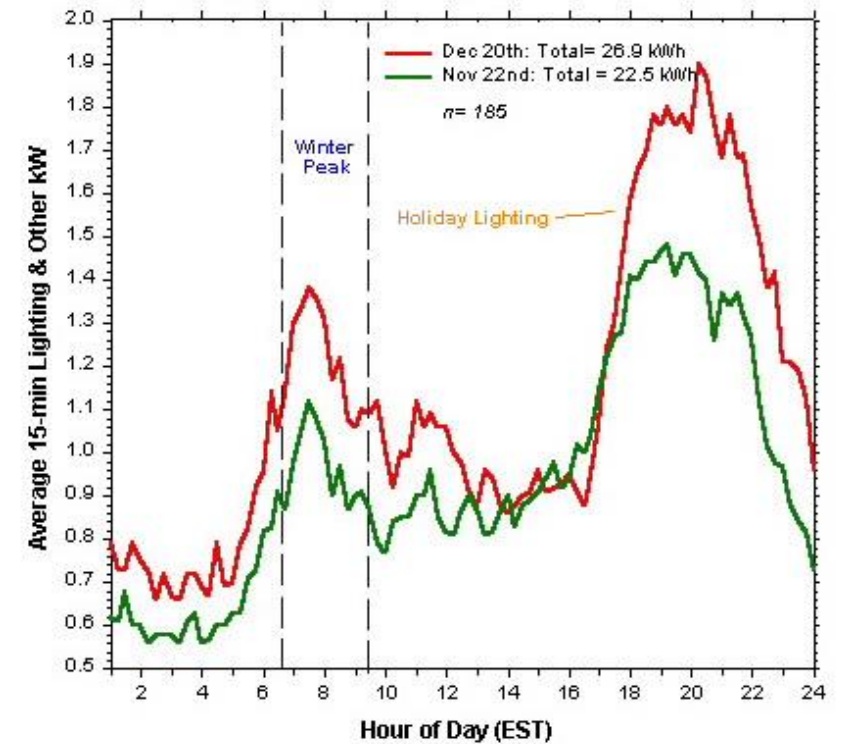
What are the inputs and outputs?

- Inputs: date, time, weather (15-minute intervals)
- Outputs: energy usage



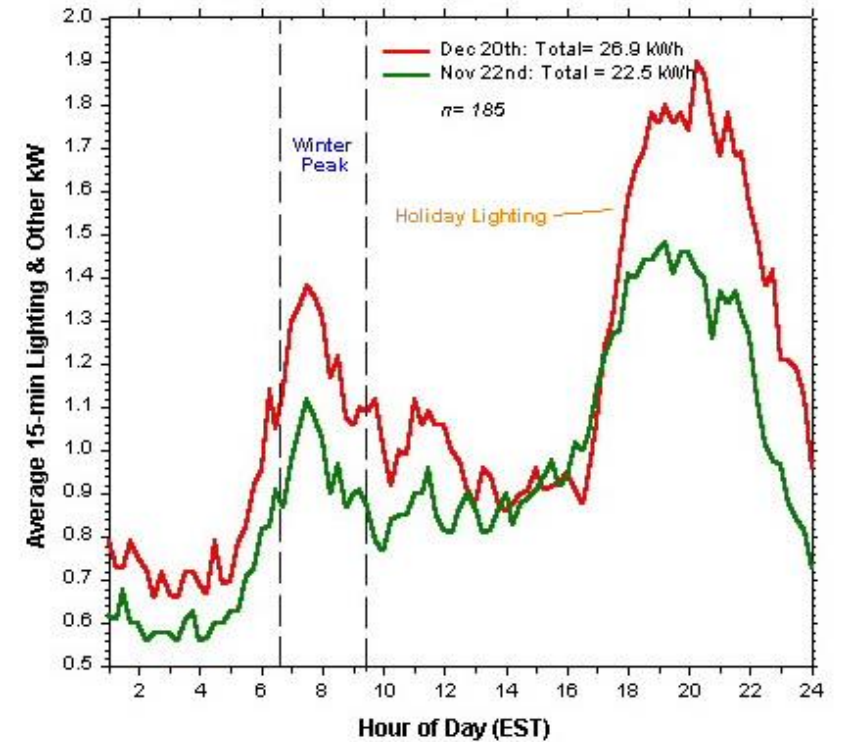
What is the labeled data?

- Previous observations



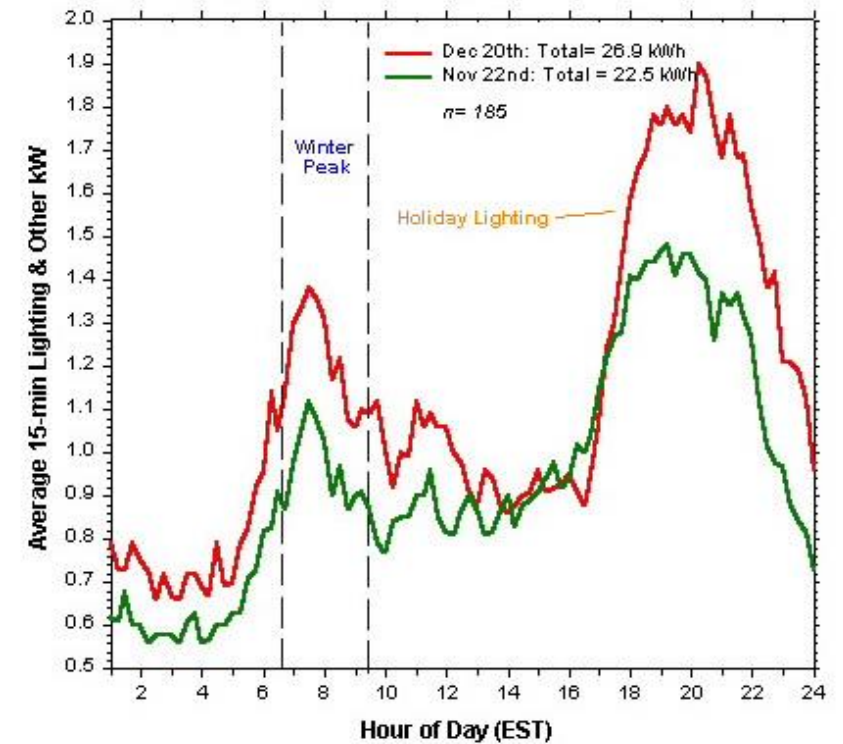
Considerations for train/test split?

- Rolling predictions



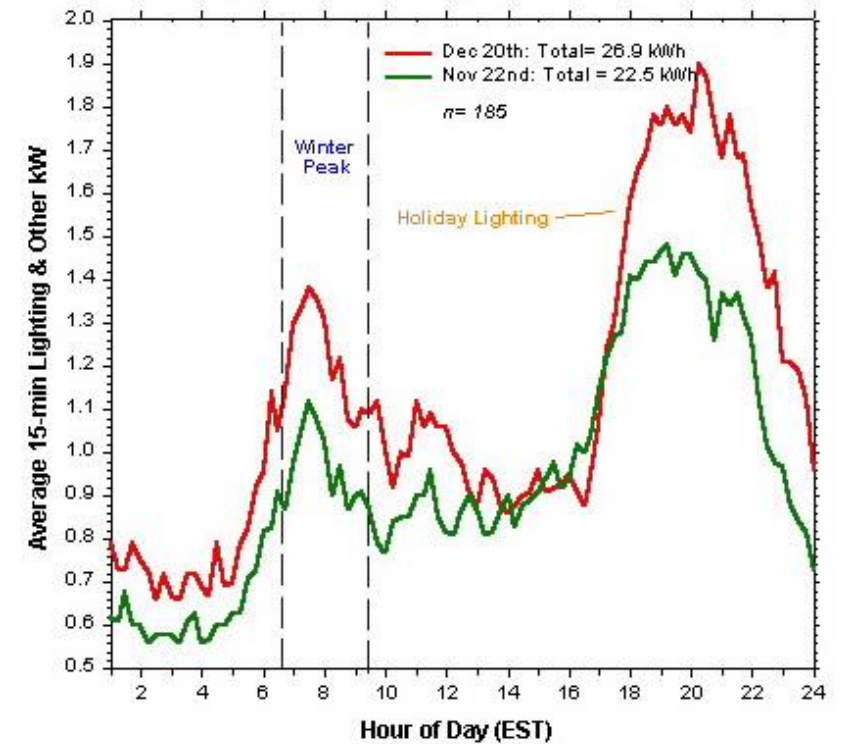
How will your function be used?

- Coordinating mixture of power sources



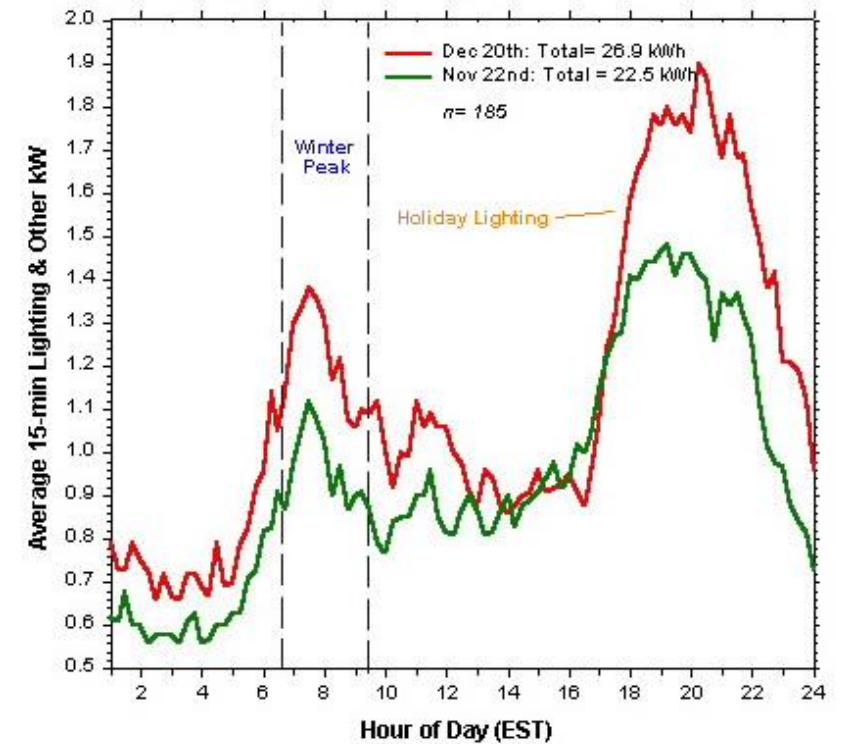
How will you evaluate predictions?

- MSE



What is a baseline predictor?

- Average usage



Review

- What is a **function**?
- What is a **model**?
- What is meant by **generalization**?
- What is **overfitting**?
- Why do we need a train/test split?
- Why do we want a baseline?

Energy Usage

The End